

Modelling

NTNU - iGem Team 2016

October 3, 2016

Abstract

The following is the documentation and analysis from the modelling of a *XOR* logic gate. This task included a formulation of the model by studying the operating principle of the gate. An expected response function was proposed accordingly. The model was solved in both dynamic state and steady state for a range of inputs for an initial set of kinetic parameters. The model was improved by modifying some equilibrium constants in a given range. An improved response surface was obtained as a result.

1 Model set up

1.1 Gate operation principle

The system we are going to model is a *XOR* gate. The behaviour for this gate is shown on Table 1.

Table 1: XOR gate table

Input		Output
A	B	A <i>XOR</i> B
0	0	0
1	0	1
0	1	1
1	1	0

However, in biological systems the inputs and outputs are not discrete values. The expected behaviour in a continuous system cannot be simplified to a table. For this purpose an ideal function has been defined:

$$z = x^2 + y^2 - 2xy \quad (1)$$

where x and y are the inputs and z is the output.

Figure 1 illustrates the behaviour of this function for values of the inputs between 0 and 1. This modelling task evaluated the behaviour of the current system compared to the ideal behaviour shown in Figure (1). Additionally, some parameters were tuned for a closer to ideal response.

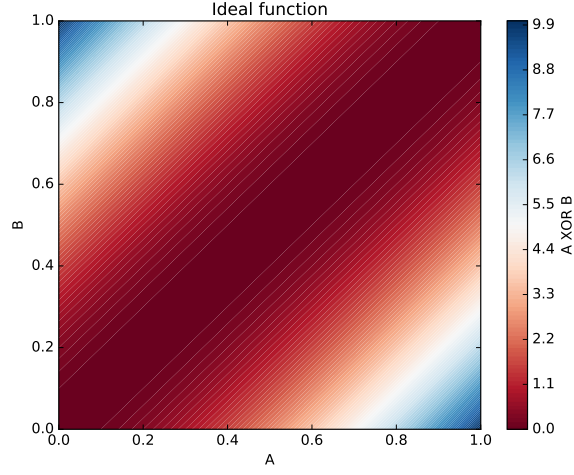


Figure 1: Ideal continuous *XOR* gate.

1.2 Chemical reactions

The biological mechanism can be summarized by several small reactions. Those reactions are occurring simultaneously. They provide the expected *XOR* gate behaviour. The main steps are shown in Figure 2.

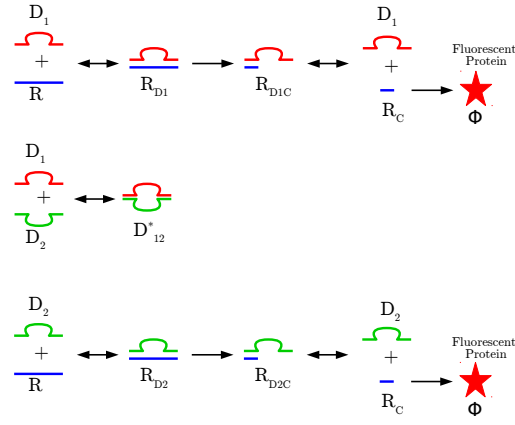


Figure 2: Main chemical reactions.

It can be seen that when both species D_1 and D_2 are present, the production of fluorescent protein is inhibited. The model should consider all of these reactions happening simultaneously. Table 2 contains the information for all the relevant reactions and their kinetic constants.

where:

- i is D_{12}^* , R_{D1} , R_{D2} , R_{D1C} , R_{D2C} and R_C .
- U_1 and U_2 are the promoters for the DNA chains (D_1 and D_2).
- R^0 : This is the production parameter for the RNA sequence. This is a constant with value of 1.

Table 2: Summary of modelled chemical reactions.

Number	Reaction	$k[s^{-1}]$	K_{eq}
1	$D_1 + D_2 \rightleftharpoons D_{12}^*$	5×10^{-3}	1400
2	$D_1 + R \rightleftharpoons R_{D1}$	5×10^{-3}	10
3	$D_2 + R \rightleftharpoons R_{D2}$	5×10^{-3}	10
4	$R_{D1} \rightarrow R_{D1C}$	8.9×10^{-3}	-
5	$R_{D2} \rightarrow R_{D2C}$	8.9×10^{-3}	-
6	$R_{D1C} \rightleftharpoons R_C + D_1$	5×10^{-1}	1000
7	$R_{D2C} \rightleftharpoons R_C + D_2$	5×10^{-1}	1000
8	$R_C \rightarrow R_c + \phi$	1.5×10^{-2}	-
9	$D_1 \rightarrow \emptyset$	1×10^{-3}	-
10	$D_2 \rightarrow \emptyset$	1×10^{-3}	-
11 (Decay)	$i \rightarrow \emptyset$	3.0×10^{-3}	-
12	$\phi \rightarrow \emptyset$	3.9×10^{-4}	-
13	$R \rightarrow \emptyset$	3×10^{-3}	-
14	$200 \rightarrow R$	9.8×10^{-2}	-
15	$U_1 \rightarrow D_1$	9.8×10^{-2}	-
16	$U_1 \rightarrow D_1$	9.8×10^{-2}	-

The inverse reaction constant for a reaction i is defined as:

$$q_i = \frac{k_i}{K_{i,eq}} \quad (2)$$

1.3 Differential equations

The following are the different differential equations for each component present in the system:

$$\frac{dD_1}{dt} = -k_1 D_1 D_2 + q_1 D_{12}^* - k_2 D_1 R + q_2 R_{D1} + k_6 R_{D1C} - q_6 R_c D_1 - k_9 D_1 + k_{15} U_1 \quad (3a)$$

$$\frac{dD_2}{dt} = -k_1 D_1 D_2 + q_2 D_{12}^* - k_3 D_2 R + q_3 R_{D2} + k_7 R_{D2C} - q_7 R_c D_2 - k_{10} D_2 + k_{16} U_2 \quad (3b)$$

$$\frac{dR}{dt} = -k_2 D_1 R + q_2 R_{D1} - k_3 D_2 R + q_3 R_{D2} - k_{13} + k_{14} R^0 \quad (3c)$$

$$\frac{dR_{D1}}{dt} = k_2 D_1 R - q_2 R_{D1} - k_4 R_{D1} - k_{11} R_{D1} \quad (3d)$$

$$\frac{dR_{D2}}{dt} = k_3 D_2 R - q_3 R_{D2} - k_5 R_{D2} - k_{11} R_{D2} \quad (3e)$$

$$\frac{dR_{D1C}}{dt} = k_4 R_{D1} - k_6 R_{D1C} + q_6 R_c D_1 - k_{11} R_{D1C} \quad (3f)$$

$$\frac{dR_{D2C}}{dt} = k_5 R_{D2} - k_7 R_{D2C} + q_7 R_c D_2 - k_{11} R_{D1C} \quad (3g)$$

$$\frac{dR_C}{dt} = k_6 R_{D1C} - q_6 R_c D_1 + k_7 R_{D2C} - q_7 R_c D_2 - k_{11} R_c \quad (3h)$$

$$\frac{d\phi}{dt} = k_8 R_c - K_{12} \phi \quad (3i)$$

$$\frac{dD_{12}^*}{dt} = k_1 D_1 D_2 - k_1 D_{12}^* - k_{11} D_{12}^* \quad (3j)$$

2 Initial case

2.1 Dynamic simulation

A dynamic simulation was carried out for this model. The following parameters were chosen:

- U_1 and $U_2 = 1$.
- $R^0 = 1$.

Figure 3 shows that a steady state is achieved. There is a large concentration of pigment given that the inputs for both signals are present.

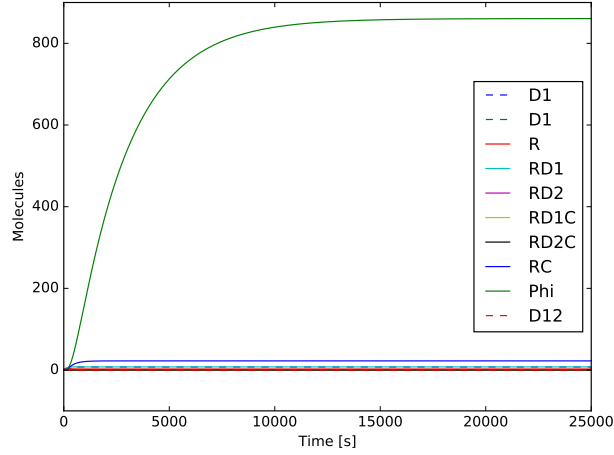


Figure 3: Dynamic response for the base case with inputs U_1 and $U_2 = 1$.

2.2 Surface response

To carry out a comparison with the ideal response, it is necessary to evaluate the system in a given surface of points.

The selected ranges for the surface are:

- $U_1 = [0, 1]$.
- $U_2 = [0, 1]$
- $R^0 = 1$.

The resulting surface is shown in Figure 4.

2.3 Observations

Figure 4 shows a trend that is similar to the ideal one. In the middle line ($U_1 = U_2$) the values for the pigment concentration are considerably smaller compared to those in the corners. However, there is a very small difference when both inputs are close to the maximum. A careful use of the current system is recommended, in order to avoid a false positive due to high values of both inputs.

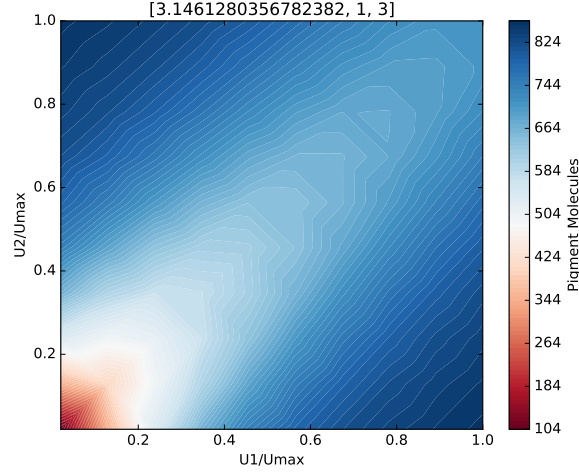


Figure 4: Surface response for the base case with input range $[0,1]$.

3 Improved case

This result shows that there is still a possibility to make improvements in the system. These are done by modifying the values of the equilibrium constants as the DNA sequences D_1 and D_2 can be designed as well as some of the RNA ones.

3.1 Parameter tuning

The constants required to be tuned are the following:

- Reaction 1: $K_{1,eq}$.
- Reaction 2,3: $K_{2,eq}$ Both reactions share the same equilibrium constant.
- Reaction 6,7: $K_{6,eq}$ Both reactions share the same equilibrium constant.

The combinations were evaluated by two criteria:

- *Similarity to the ideal function.* This means that the trend should be as close as the ideal as possible. The situation where both components are present should be zero.
- *Amount of pigment particles produced.* This ensures that the trend is kept for a significant amount of pigment particles. It is possible for some surface to have a behaviour close to ideal while producing very few pigment particles.

The implementation of these two criteria into a single function for an optimization problem did not meet the requirements. Thus, a manual search over a space was carried out. The parameters were evaluated in a logarithmic scale. The following ranges were used:

Table 3: Equilibrium constants tuning ranges.

Constant	Lower Bound	Upper Bound
$K_{1,eq}$	10^2	10^6
$K_{2,eq}$	10^{-4}	10^6
$K_{6,eq}$	10^{-4}	10^6

Table 4: Final equilibrium constants values.

Constant	Value
$K_{1,eq}$	10^5
$K_{2,eq}$	$10^{-1.5}$
$K_{6,eq}$	10^3

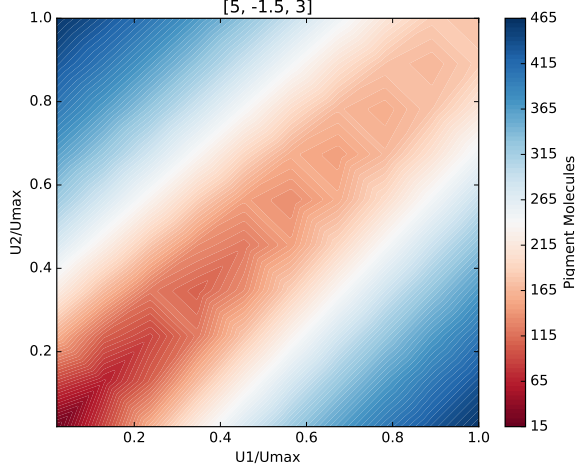


Figure 5: Optimal surface response for the base case with input range $[0,1]$.

3.2 Results

Searching in the ranges summarized on Table 3, the obtained values for the equilibrium constants are shown on Table 4. Figure 5 illustrates the resulting surface. Figure 5 shows a response very close to ideal. It has a very sharp difference for the case when there is only one input (between 350 and 465 pigment molecules) compared to the simultaneous inputs (15 to 165). Additionally, there is a significant amount of pigment molecules produced for when the gate response is positive (more than 300).

The modifications on the constants have the following implications on the reactions:

- $K_{1,eq} = 10^5$. The component D_{12}^* should be very stable and formed quickly. This ensures that the pigment production is inhibited while both molecules are present.
- $K_{2,eq} = 10^{-1.5}$. The first RNA binding (reactions 2 and 3) should be slow enough to allow both molecules D_1 and D_2 to bind if they are present blocking the pigment production. If this reaction is too fast the inhibition reaction will not have a significant impact on the pigment production.
- $K_{6,eq} = 10^3$. Reactions (6 and 7) should be fast. This allows for a quicker production of pigment and amplifies the effect from the previous equilibrium reactions.

4 Conclusions

It was possible to model the *XOR* logic gate system, both dynamically and steady state for a series of input.

An expected response surface was computed and it shows a similar response to the ideal one using the current information.

The surface was improved by modifying the equilibrium constants in the model. This provided valuable feedback on the reactions that need to be promoted and those that need to be inhibited.